



Restaurant Review Classifier

Sentiment Analysis for Restaurant Feedback

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Problem Statement & Objectives

Problem Statement

Manually analyzing large volumes of customer feedback is time-consuming and can miss valuable insights.



Objectives

- Classify reviews as Positive or Negative
- Extract key aspects (Food Quality, Service, Hygiene, Staff)
- Provide summarized insights for decision-making
- Improve brand monitoring and customer satisfaction



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Workflow

- [illegible]

Dataset & Methodology



Dataset Used

Restaurant_Reviews.tsv: Contains review text and sentiment labels for model training and testing.



Data Preprocessing

Using NLTK for cleaning (punctuation, numbers, stopwords), tokenization, and stemming.



Feature Extraction

TF-IDF Vectorization to convert text into numerical features.



Model Training

Naive Bayes algorithm for sentiment classification.



Deployment

FastAPI backend with a Jinja2 template frontend.

Technologies Used



Frontend

HTML,CSS,Jinja2 Templates



Backend

FastAPI, Uvicorn



Programming Language

Python



ML / NLP Tools

Scikit-Learn,NaiveBayes, TF-IDF, NLTK



Storage

Pickle(.pklmodel & vectorizer)



Dataset

Restaurant_Reviews.tsv

Results & Findings

Good Accuracy

Naive Bayes model achieved strong performance on the dataset.

Improved Features

TF-IDF significantly enhanced feature quality for better prediction.

Effective Classification

System successfully distinguishes Positive and Negative reviews.

Real-time Prediction

Fast API deployment enables instant sentiment analysis.

Quick Insights

Helps restaurants rapidly understand customer satisfaction.

Conclusion & Future Scope

Conclusion

The RestaurantReview Classifier effectively automates sentiment prediction using NLP and ML, offering a lightweight, fast, and deployable solution via FastAPI.



Future Scope

- Full analytics dashboard
- Multi-language support
- Aspect-based analysis (Food, Service, Ambience)
- Upgrade to transformer models (BERT, RoBERTa)
- Cloud deployment (AWS / Render / Azure)

